

No. data points : n

No. labels : L

Micro averaging F-measure

$$\hat{TP}_{\text{micro}}(f) := \frac{1}{nL} \sum_{i=1}^n \sum_{j=1}^L \mathbb{I}\{f_j(x_i) = 1, y_j^i = 1\}$$

$$\hat{TN}_{\text{micro}}(f) := \frac{1}{nL} \sum_{i=1}^n \sum_{j=1}^L \mathbb{I}\{f_j(x_i) = 0, y_j^i = 0\}$$

$$\hat{\Theta}_{\text{micro}} := \frac{\sum_{i=1}^n \sum_{j=1}^L \mathbb{I}\{y_j^i = 0\}}{\sum_{i=1}^n \sum_{j=1}^L \mathbb{I}\{y_j^i = 1\}}$$

$$\hat{F}_{\beta}^{\text{micro}}(f) := \frac{(1+\beta^2) \cdot \hat{TP}_{\text{micro}}(f)}{\beta^2 + \hat{\Theta}_{\text{micro}} + \hat{TP}_{\text{micro}}(f) - \hat{\Theta}_{\text{micro}} \cdot \hat{TN}_{\text{micro}}(f)}$$

$$\hat{F}_{\beta}^{\text{micro}}(f) \geq v \iff (1+\beta^2-v) \cdot \hat{TP}_{\text{micro}}(f) + v \cdot \hat{TN}_{\text{micro}}(f) \geq v(\beta^2 + \hat{\Theta}_{\text{micro}})$$

AMP_{micro}

$$v_{t+1} \leftarrow \hat{F}_{\beta}^{\text{micro}}(f_t)$$

$$f_{t+1} \leftarrow \underset{f \in \mathcal{F}}{\text{argmax}} \left\{ \underbrace{(1+\beta^2-v_{t+1}) \cdot \hat{TP}_{\text{micro}}(f)}_{a_{t+1}} + \underbrace{v_{t+1} \cdot \hat{TN}_{\text{micro}}(f)}_{b_{t+1}} \right\}$$

$$a_{t+1} \hat{TP}_{\text{micro}}(f) + b_{t+1} \hat{TN}_{\text{micro}}(f) = \frac{1}{nL} \sum_{i=1}^n \sum_{j=1}^L (a_{t+1} \hat{TP}_j^i(f) + b_{t+1} \hat{TN}_j^i(f))$$

Define

$$\begin{cases} \hat{TP}_j^i(f) = \mathbb{I}\{f_j(x_i) = 1, y_j^i = 1\} \\ \hat{TN}_j^i(f) = \mathbb{I}\{f_j(x_i) = 0, y_j^i = 0\} \end{cases}$$

$$= \frac{1}{L} \sum_{j=1}^L \frac{1}{n} \sum_{i=1}^n a_{t+1} \hat{TP}_j^i(f_j) + b_{t+1} \hat{TN}_j^i(f_j)$$

$$f_{t+1} = \{f_{t+1}^1, f_{t+1}^2, f_{t+1}^3, \dots, f_{t+1}^L\}$$

f_{t+1}^j is obtained by solving a ^{binary} weighted classification problem on the j^{th} label points with the costs a_{t+1} and b_{t+1} on positives and negatives respectively.

$$\begin{aligned} a_{t+1} &= 1 + \beta^2 - v_{t+1} \\ b_{t+1} &= v_{t+1} \cdot \hat{\Theta}_{\text{micro}} \end{aligned}$$

Macro averaging F-measure

$$\hat{T}P_{\text{macro}}^j(f) = \frac{1}{n} \sum_{i=1}^n \hat{T}P_i^j(f)$$

$$\hat{T}N_{\text{macro}}^j(f) = \frac{1}{n} \sum_{i=1}^n \hat{T}N_i^j(f)$$

$$\hat{\Theta}_{\text{macro}}^j = \frac{\sum_{i=1}^n \mathbb{I}\{y_i^j = 0\}}{\sum_{i=1}^n \mathbb{I}\{y_i^j = 1\}}$$

$$\hat{F}_{\beta}^{\text{macro}}(f) = \frac{1}{L} \sum_{j=1}^L \hat{F}_{\beta}^j(f^j)$$

$$\hat{F}_{\beta}^j(f^j) = \frac{(1+\beta^2) \hat{T}P_{\text{macro}}^j(f^j)}{\beta^2 + \hat{\Theta}_{\text{macro}}^j + \hat{T}P_{\text{macro}}^j(f) - \hat{\Theta}_{\text{macro}}^j \cdot \hat{T}N_{\text{macro}}^j(f)}$$

Similarly as before, $\hat{F}_{\beta}^j(f^j) \geq v^j \Leftrightarrow \underbrace{(1+\beta^2 - v^j) \cdot \hat{T}P_{\text{macro}}^j(f^j)}_{\star} + v^j \cdot \hat{T}N_{\text{macro}}^j(f) \geq v^j (\beta^2 + \hat{\Theta}_{\text{macro}}^j)$

AMP_{macro}

$$v_{t+1}^j \leftarrow \hat{F}_{\beta}^j(f_t^j)$$

$$f_{t+1}^j \leftarrow \underset{f^j}{\text{argmax}} \{ a_{t+1}^j \hat{T}P_{\text{macro}}^j(f^j) + b_{t+1}^j \hat{T}N_{\text{macro}}^j(f^j) \}$$

$$\boxed{\begin{aligned} a_t^j &= (1+\beta^2 - v_t^j) \\ b_t^j &= v_t^j \cdot \hat{\Theta}_{\text{macro}}^j \end{aligned}}$$

$$f_{t+1} = \{f_{t+1}^1, f_{t+1}^2, \dots, f_{t+1}^L\}$$

f_{t+1}^j is solved by obtaining a cost weighted classification problem with the costs a_t^j and b_t^j on the positives (points that have the label j) and negatives (points which do not have label j)

Minimizing the instance weighted F-measure on data is an NP-hard problem in general.